

Experts in uncertainty

An introduction to the classical model

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The classical model



- The classical method
- Cooke's model/method
- Cooke's classical model/method
- Classical model for expert judgement
- Structured expert judgment (SEJ)
- Structured expert judgment: the classical model (wikipedia)
- Structured elicitation of expert judgement(s)

Why EJ?

- No (or not much) data available
- Data sources in an inadequate form for the analysis
- Data sources fraught with problems (i.e., poor entry, bad data definitions, etc.)

SEJ elicitation

- The process (protocol) of obtaining information about uncertain events from experts
- Variable of interest

When will man land on Mars?

5% _____, 50% _____, 95% _____

SEJ elicitation

- The process (protocol) of obtaining information about an uncertain events (probability distributions) from experts
- Variable of interest

When will man land on Mars?

e_1 : 5% 2035, 50% 2050, 95% 2070

e_2 : 5% 2018, 50% 2025, 95% 2100

- 50% quantile (50th percentile) – best guess
- 5% quantile (5th percentile) – $P(X \leq 2035) = 0.05$
- 95% quantile (95th percentile) – $P(X \geq 2070) = 0.05$

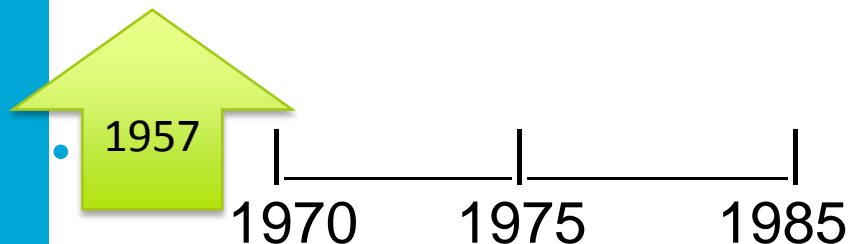
SEJ elicitation

- Questions for which we know the true values (calibration, performance, seed questions/variables)

What was the 1946 RAND forecast for year of first launched satellite?

e_1 : 5% 1970, 50% 1975, 95% 1985

e_2 : 5% 1950, 50% 1960, 95% 2000

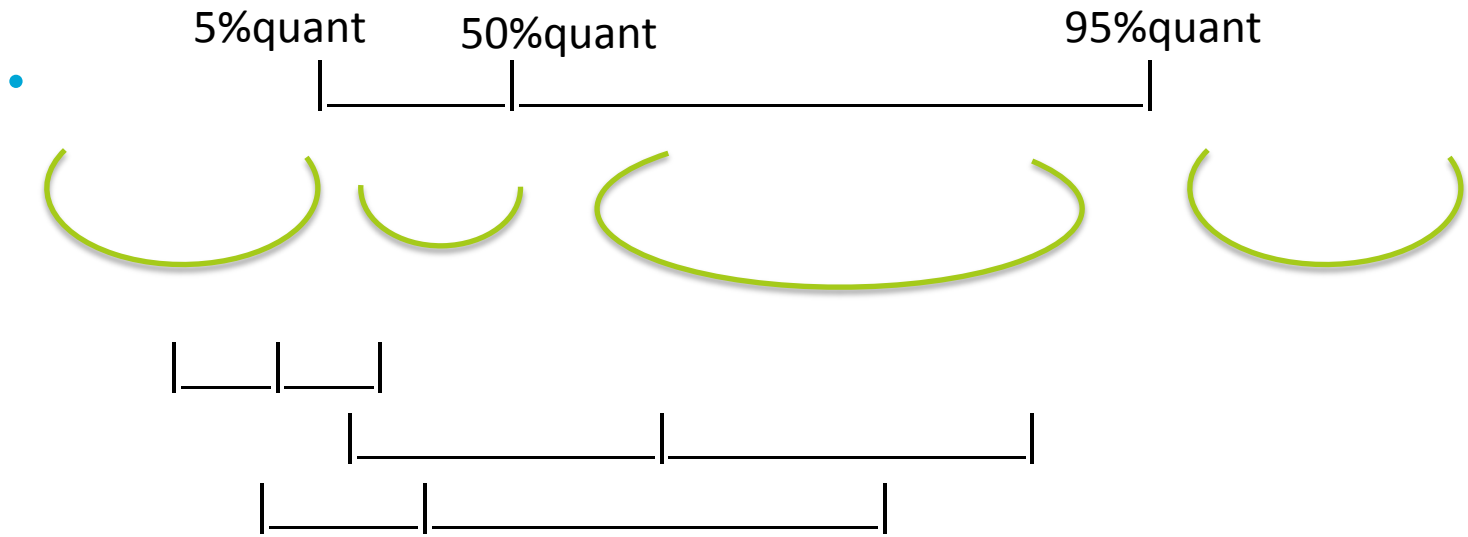


- Sputnik – 4 October, 1957

Calibration

- Are the expert's probability statements statistically accurate?
- Calibration measures how well experts' assessments correspond to actual values (realizations)
- Statistical accuracy
- Calibration is a statistical question – we need a sufficient number of calibration variables (seed questions) to compare assessments to realizations with any confidence

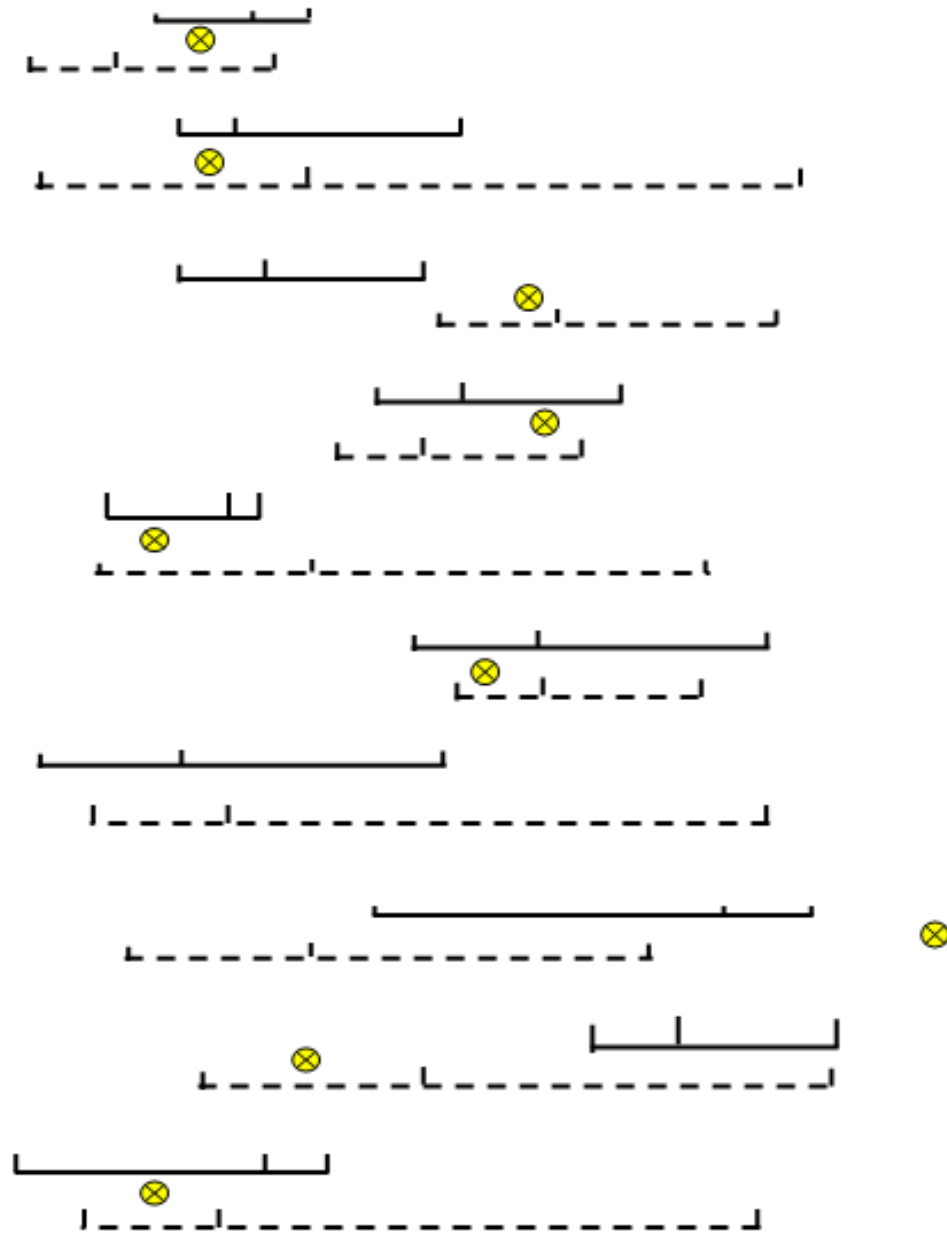
Calibration



- $p_1 = 0.05, p_2 = 0.45, p_3 = 0.45, p_4 = 0.05$
- $p = (p_1, p_2, p_3, p_4)$
- $s = (s_1, s_2, s_3, s_4)$ for expert e_1 ?

Calibration

- $s_1=0.2, s_2=0.5$
 $s_3 = 0.1, s_4=0.2$
(for expert 1)
- $s_1=0.1, s_2=0.6$
 $s_3 = 0.2, s_4=0.1$
(for expert 2)



Calibration

- We expect some discrepancy between s and p , due to random fluctuation
- To measure the discrepancy between s and p , we use the relative information of s relative to p

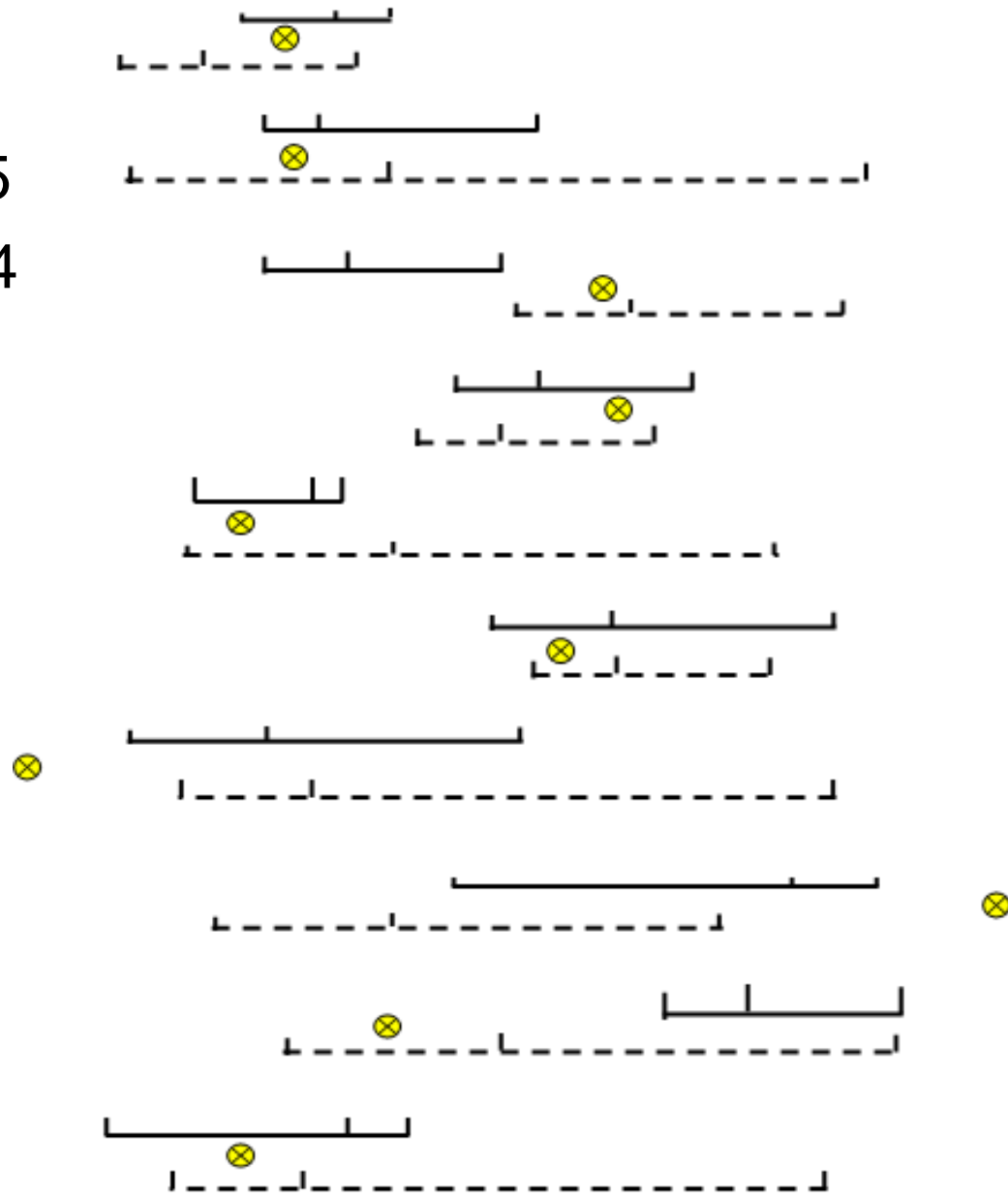
$$I(s, p) = \sum_{i=1}^4 s_i \ln(s_i/p_i)$$

- For n questions and 3 elicited quantiles

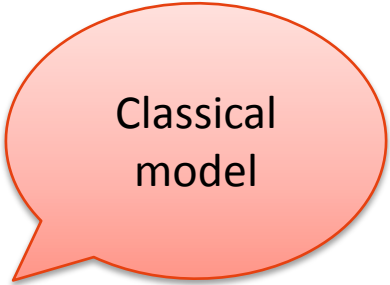
$$\text{Cal}(e) = 1 - \chi^2_3(2nI(s, p))$$

Calibration

- $\text{Cal}(1)=0.0275$
 $\text{Cal}(2)=0.3944$



Calibration



Classical
model

- Measure the degree to which the data supports the hypothesis that the expert's probabilities are accurate
- We do not say we (fail to) reject expert hypothesis
- Low scores, near zero, mean that it is unlikely that the expert's probabilities are correct
- High scores indicate good support

Information

- Information in a distribution is the degree to which the distribution is concentrated

What was the 1946 RAND forecast for
year of first satellite (1957)?

e_1 : 5% 1960, 50% 1965, 95% 1970

e_2 : 5% 1950, 50% 1960, 95% 2000

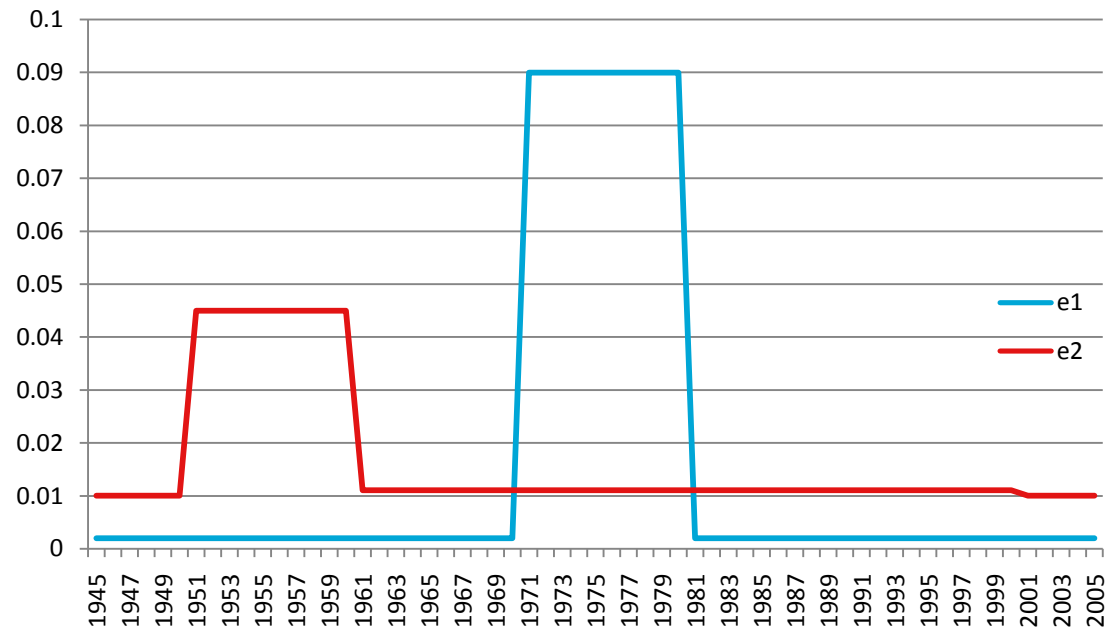
- Measuring information requires associating a density with each assessment of each expert

Information

- Information is measured with respect to a background measure (chosen by the analyst)
 - Uniform
 - Log-uniform
- The background measures require an *intrinsic range* on which measures are concentrated
- $k\%$ overshoot rule
 - For each variable, the smallest interval $I=[L,U]$
 - Extend I to $I^* = [L^*,U^*]$, where
$$L^* = L - \frac{k(U - L)}{100}, U^* = U + \frac{k(U - L)}{100}$$
- k is chosen by the analyst (typically $k=10$)

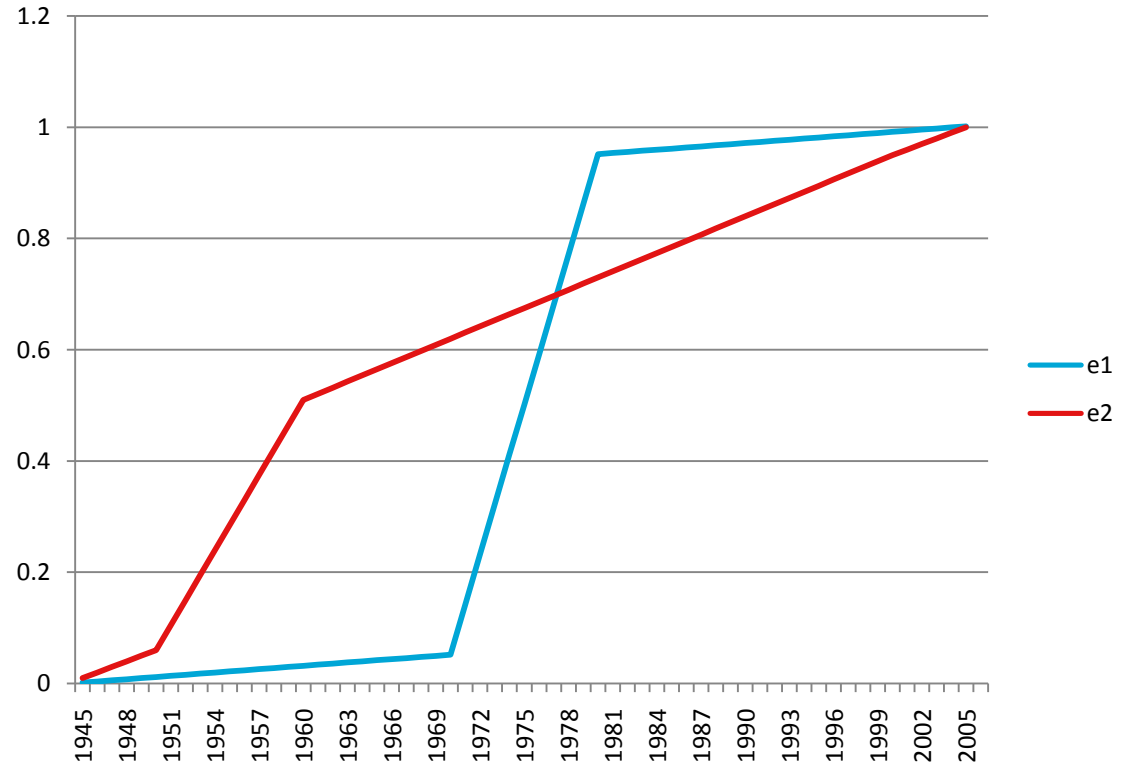
Information

- Probability mass function which agrees with experts' percentile assessments, relative to the uniform measure



- $I^* = [1945, 2005]$

Information



Information

- $\text{Inf}(i, \text{Unif}) =$ relative information wrt background

$$\text{Inf}(e) = \frac{1}{n} \sum_{i=1}^n \text{Inf}(f(i), \text{Unif})$$

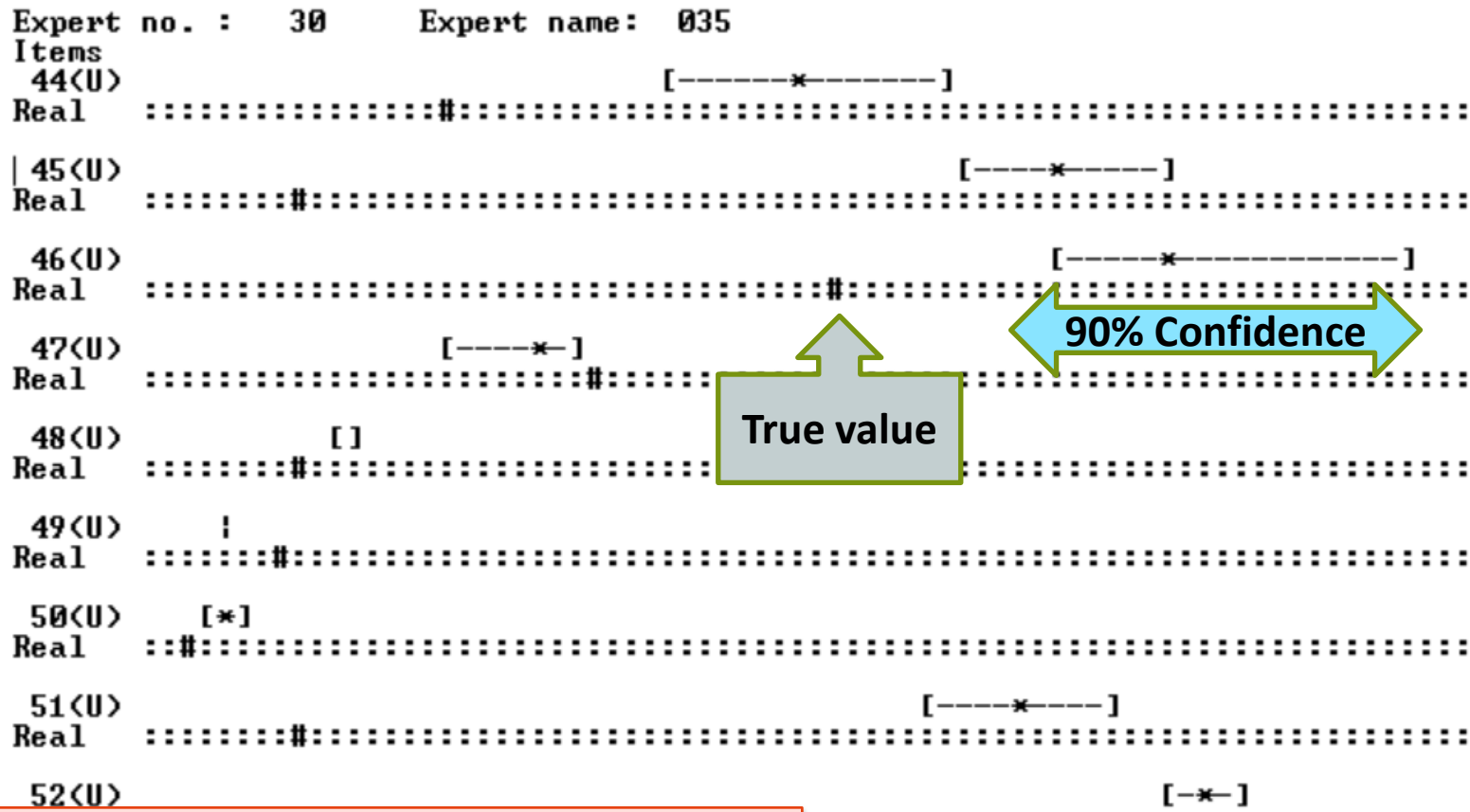
- High values indicate that the expert is adding “a large amount of information” to the background distribution

What is a GOOD subjective probability assessor?

- **Calibration (statistical accuracy)**
 - Are expert's probability statements statistically accurate?
 - A high calibration score
- **Informativeness**
 - Probability mass concentrated in a small region, relative to the background measure
 - High information score

Expert performance CAN be objectively measured

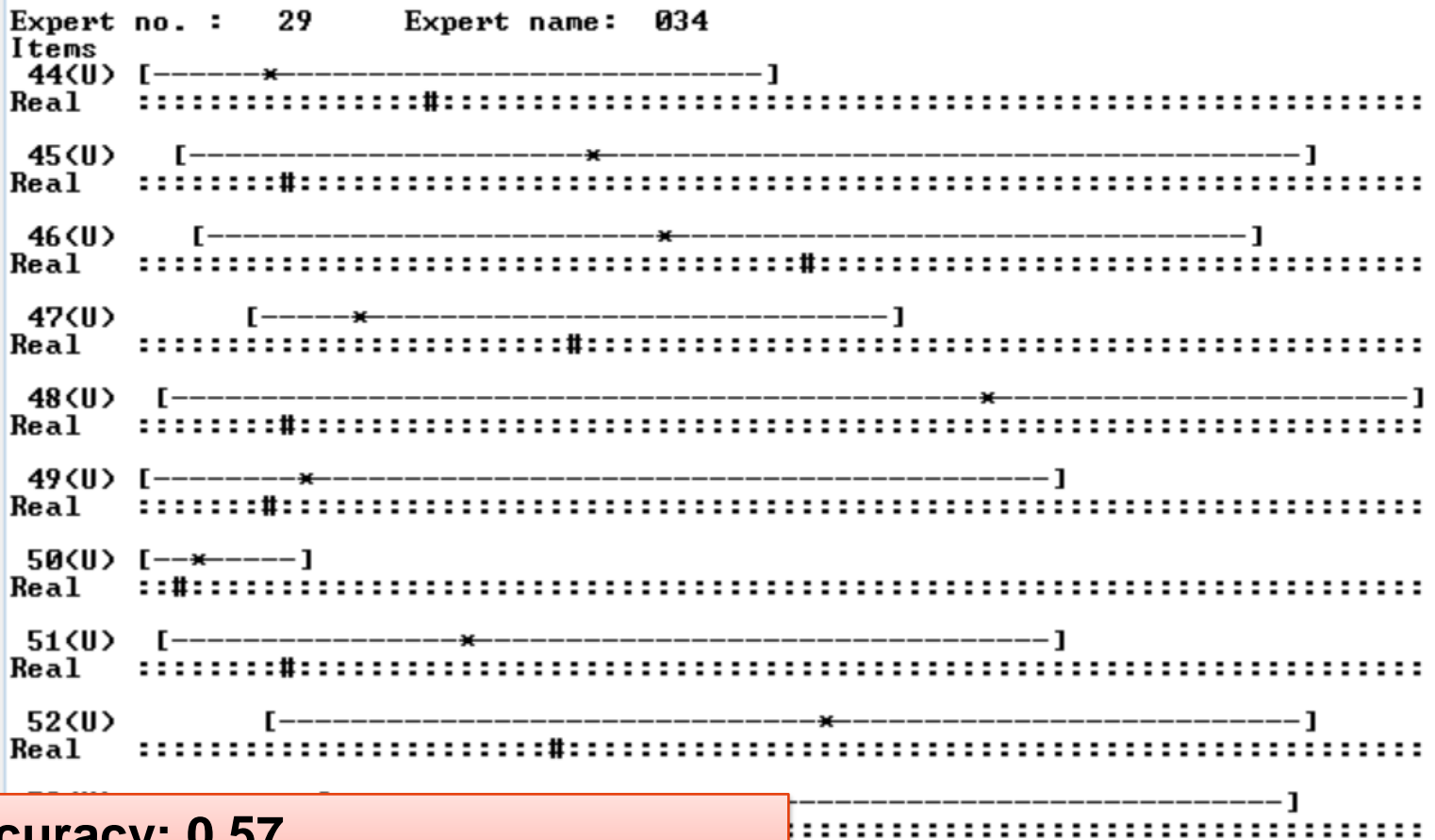
Very High Information, Very Poor Statistical Accuracy



Statistical Accuracy: 0.00000067
Informativeness: 2.282

Expert performance CAN be objectively measured

Low Information, Good Statistical Accuracy



Statistical Accuracy: 0.57
Informativeness: 0.53

Expert performance CAN be objectively measured

High Information, Decent Statistical Accuracy

```
Expert no. : 16      Expert name: 018
Items
44(U)      [-----*-----]
Real .....#.....

45(U)      [--*--]
Real .....#.....

46(U)      [-----*-----]
Real .....#.....

47(U)      [---*---]
Real .....#.....

48(U)      [-*--]
Real .....#.....

49(U)      [-*-]
Real .....#.....

50(U)      [--*]
Real .....#.....

51(U)      [--*-]
Real .....#.....

52(U)      [-----*-----]
Real .....#.....

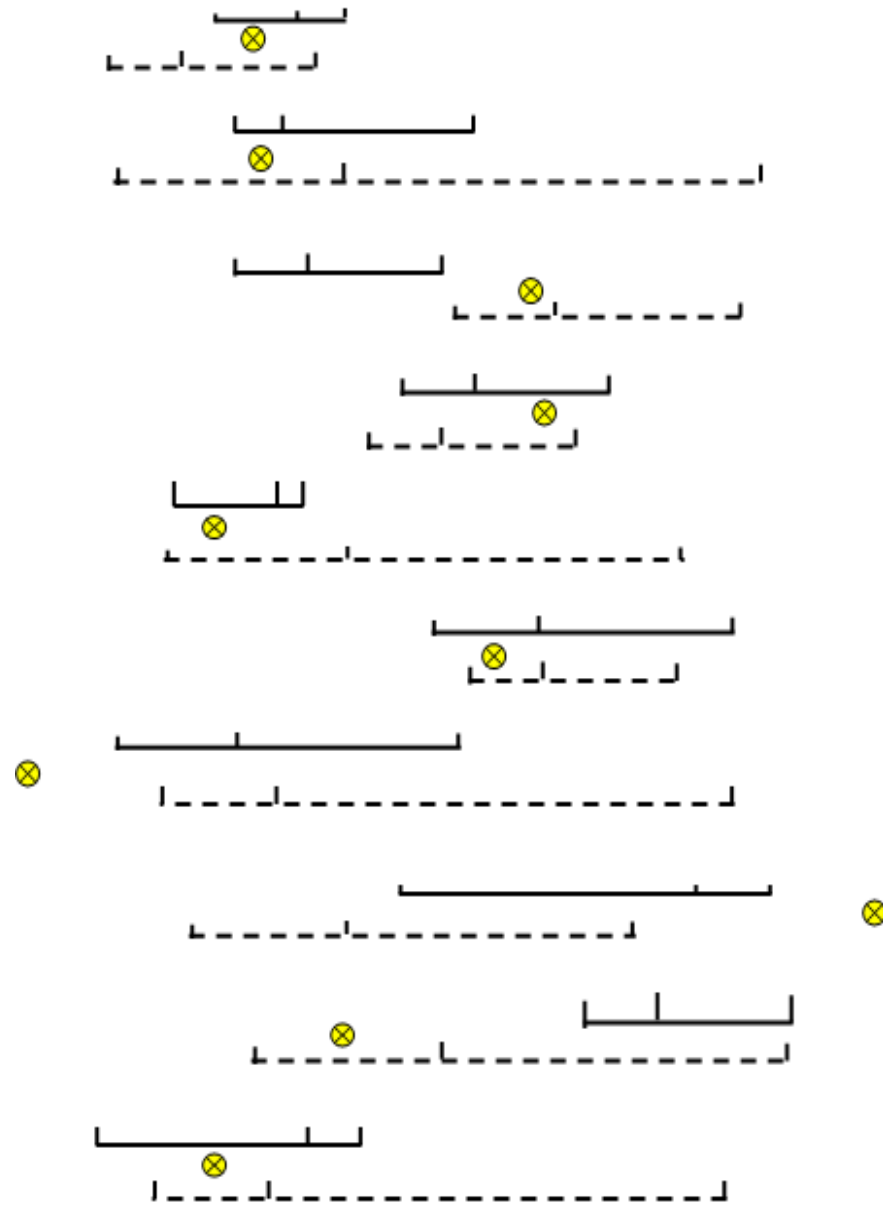
53(U)      [-----*-----]
Real .....#.....
```

Statistical Accuracy: 0.12

Informativeness: 1.77

Combined score

- **Cal(e)*Inf(e)**
- **Score(1)=**
 $0.0275 * 1.149 = 0.0315$
- **Score(2)=**
 $0.3944 * 0.5912 = 0.2331$



Decision Maker (DM)

Expert	Calibration	Informativeness
Expert 1	0.06083	0.91
Expert 2	0.00628	1.56
Expert 3	0.01397	1.24
Expert 4	0.6827	0.82
Expert 5	0.002809	1.15
Expert 6	0.05706	1.32
Expert 7	0.01397	1.10

- $\text{Cal}(e) \geq 0.05$

Combined expert score

Significance Level

- $w_{\alpha}(e) = Cal(e) * Inf(e) * \underbrace{1\{Cal(e) \geq \alpha\}}$

= 1 if calibration $\geq \alpha$, else = 0

- This score is an asymptotically proper scoring rule, i.e.
 - Expert maximizes long run expected score by, and only by, stating percentiles which (s)he believes

$$DM_{\alpha} = \frac{\sum_{i=1}^E w_{\alpha}(e_i) f_{e,i}}{\sum_{i=1}^E w_{\alpha}(e_i)}$$

Optimization

-
- Q: How to choose α ?
- A: For each α , compute

$$Cal(DM_\alpha) * Inf(DM_\alpha)$$

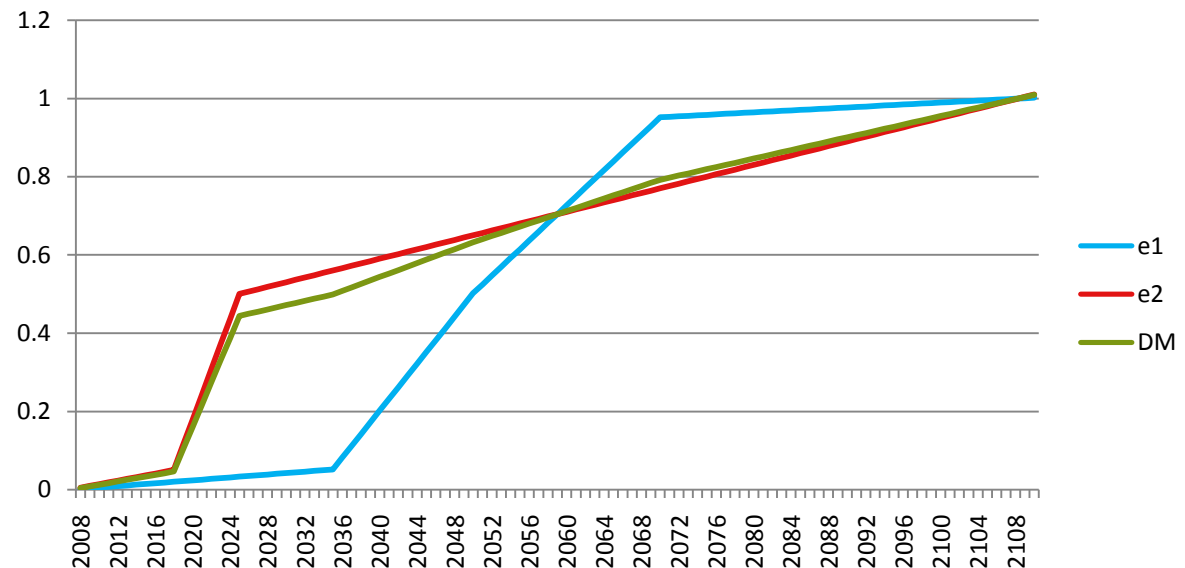
- choose α for which this is maximum

Decision Maker (DM)

- **Performance based weights (PW)**

$$w_1 = \frac{0.0315}{0.0315 + 0.2331} = 0.12$$

$$w_2 = \frac{0.2331}{0.0315 + 0.2331} = 0.88$$



Another approach – Averaging quantiles

- **Performance based weights (PW)**

$$w_1 = 0.12$$

$$w_2 = 0.88$$

- **$DM_{PW} = w_1 e_1 + w_2 e_2$**

- 50% - $0.12 \cdot 2050 + 0.88 \cdot 2025 = 2028$

- 5% - $0.12 \cdot 2035 + 0.88 \cdot 2018 = 2020$

- 95% - ...

Another option - Equal weights

- When will man land on Mars?

e_1 : 5% 2035, 50% 2050, 95% 2070

e_2 : 5% 2018, 50% 2025, 95% 2100

- Equal weights?

– $50\% - \frac{1}{2} 2050 + \frac{1}{2} 2025 = 2037$

– 5% ..., 95% ..., $f(i)$

$$DM_{EW} = \frac{1}{2}e_1 + \frac{1}{2}e_2$$

Combining experts

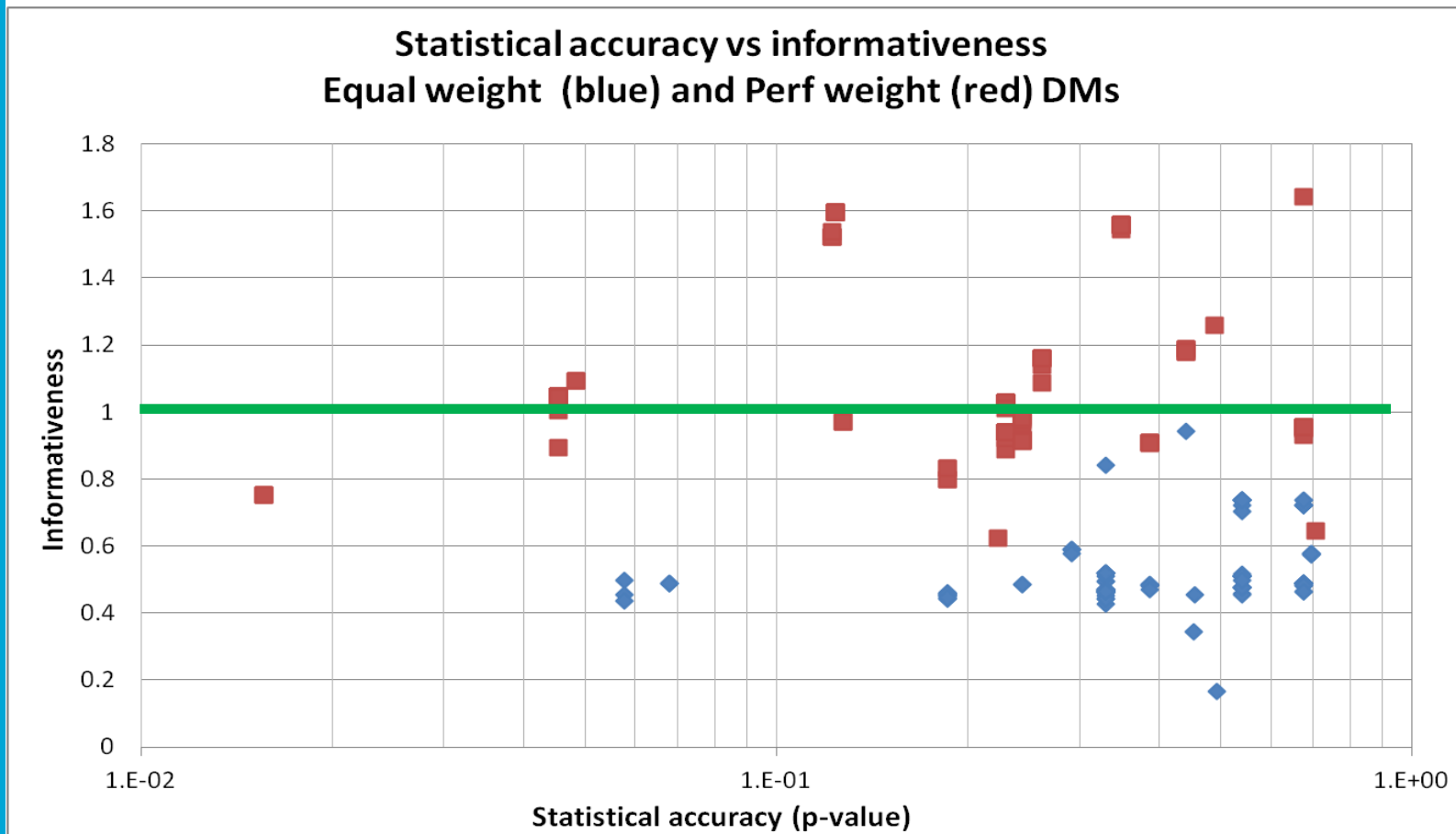
- Equal weights

Performance based

- Global weights: $\text{Calibr} * \text{Ave Inf} * \text{cutoff}$
Item weights: $\text{Calibr} * \text{Inf per item} * \text{cutoff}$

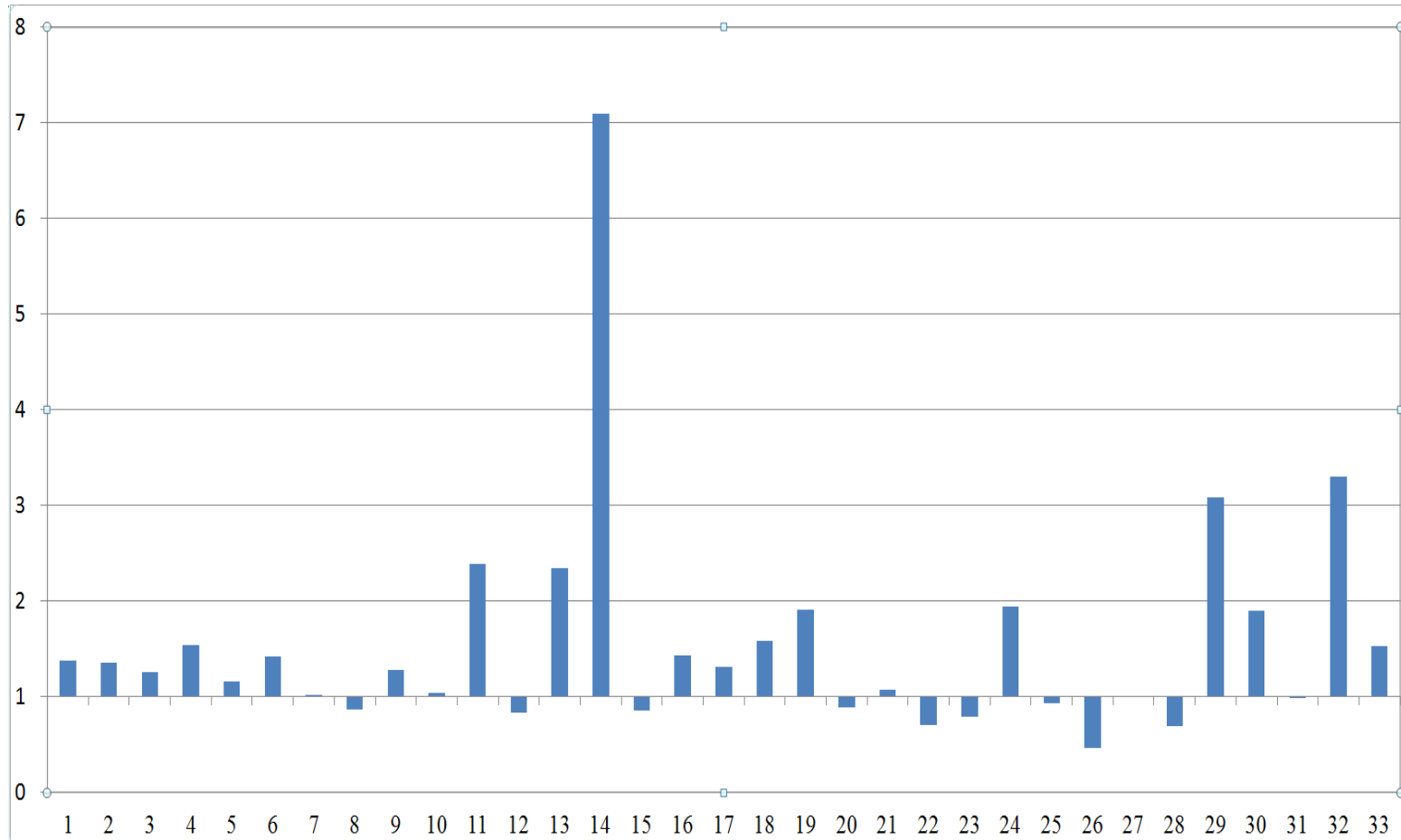
Why bother?

- **Performance based combination of experts improves statistical accuracy and informativeness**



Out of sample validation

- Training sets sized at 80% of the calibration variables
- Ratio PWCombined/EWCombined



Questions?



SEJ for rational consensus

- Parties pre-commit to a method which satisfies necessary conditions for scientific method
 - Traceability/accountability
 - Neutrality (don't encourage untruthfulness)
 - Fairness (ab initio, all experts are equal)
 - Empirical control (performance measurement)
- Goal: comply with the principles and combine experts' judgements to get a GOOD probability assessor

SEJ theses

- SEJ is not knowledge
- Experts can quantify uncertainty as subjective probability
- Experts don't agree
- We can do better than equal weighting
- *The choice is not whether to use expert judgment, but whether to do it well or badly*